

# Enabling the Usability of Earth Science Data Products and Services by Evaluating, Describing, and Improving Data Quality throughout the Data Lifecycle

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*Robert R. Downs<sup>1</sup>, Ge Peng<sup>2</sup>, Yaxing Wei<sup>3</sup>,  
Hampapuram Ramapriyan<sup>4</sup>, and David F Moroni<sup>5</sup>*

<sup>1</sup> *rdowns@ciesin.columbia.edu*

NASA Socioeconomic Data and Applications Center (SEDAC), CIESIN, Columbia University

<sup>2</sup> *ge.peng@noaa.gov*

NC State University, Asheville, NC,

<sup>3</sup> *weiy@ornl.gov* Oak Ridge

National Laboratory, Oak Ridge, TN,

<sup>4</sup> *hampapuram.ramapriya@ssaihq.com*

Science Systems and Applications, Inc., Lanham, MD,

<sup>5</sup> *david.f.moroni@jpl.nasa.gov*

NASA Jet Propulsion Laboratory, Pasadena, CA

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# Overview



- What is data quality and why do we need data quality assessments?
- Data quality in standards and guidelines
- Data quality in data lifecycle and workflows
- Identified actions and implementations for improving data quality
- Implications and progressive planning for data quality assessments
- Recommendations and taking action

# What is data quality and why is it important?

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- Assessment of the potential usefulness of data and metadata
  - Data quality can be assessed for multiple purposes
  - Metadata often includes information on data quality characteristics to enhance overall data quality through self-description
- Data quality offers value to enable use
  - Quality attained for internal use might be insufficient for external use
  - Enabling public use of data may require higher levels of data quality
  - Investments in data curation become worthwhile when data are used
- Potential users need to determine potential usefulness of data
  - An assessment of data quality is necessary to determine usefulness
  - Each potential use of data could require a different assessment

# Why do we need data quality assessments?

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- Identify potential opportunities for users to use the data
  - Usefulness for a particular purpose, including science studies, education, and decision-making
  - Limitations of methods, variables, or values
- Types of assessments can reflect a variety of potential uses
  - Interdisciplinary use could require multiple assessments
  - New uses of data may require new assessments
- Improve trustworthiness of the data product
  - Increase science transparency by describing quality and limitations
  - Provide independent review of data
- Describe potential for data use beyond the initial data study
  - Indicate opportunities for subsequent use
  - Identify opportunities to combine data with other data

# Data Quality in Standards and Guidelines

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- Federal Geographic Data Committee (FGDC) Content Standard for Digital Geospatial Metadata (CSDGM) FGDC-STD-001-1998
  - Report on the data quality assessment
- ISO 19115-1:2014 Geographic information -- Metadata -- Part 1: Fundamentals
- ISO 19157:2013 Geographic information -- Data quality
  - Describing, registering, evaluating, and reporting data quality
- NASA Earth Science Data Preservation Content Specification (PCS). 2013.
  - Product quality, including methods, quality flags, uncertainty, and limitations
- ISO 14721:2012 Space data and information transfer systems -- Open archival information system (OAIS) -- Reference model.
  - Data quality reviews are reported in Representation Information
- GEO Data Management Principles Implementation Guidelines
  - Traceability, Data Quality Control, and Data Review and Reprocessing

# Data Lifecycle Contributions to Data Quality

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## Data users

Data use team, funders, institutions, and reviewers identify and describe data quality for their uses



## Data disseminators

Disseminators, intermediaries, funders, institutions and reviewers enable discovery and use of data quality information for users



## Data curators

Data curation team, funders, institutions and reviewers evaluate and document quality of data for potential future uses envisioned



## Data collectors

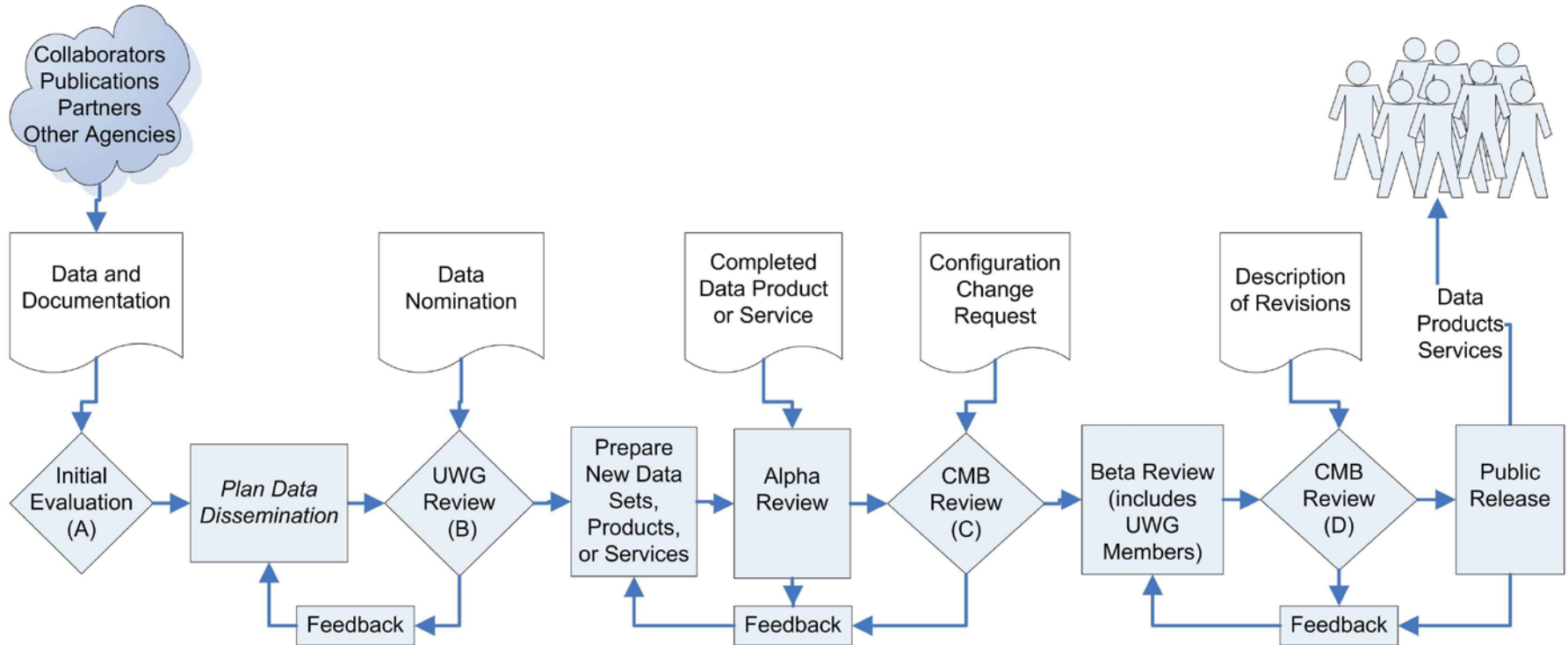
Science team, their funders, institutions and reviewers identify and document data quality issues for the intended uses

# Data Quality in Scientific Data Study Workflow



- Conceptualize study
  - Identify and document potential uses of data
- Design study
  - Describe potential uses of data and data review and documentation process
- Collect data
  - Document data collection instruments, variables, procedures, and anomalies
- Analyze data
  - Document assumptions about data for analysis and limitations
- Deposit data
  - Submit data quality information with data
- Publish results
  - Describe limitations of data use and cite data and data quality information

# Data Quality Workflow in Domain Repository



Example: NASA SEDAC data review process with feedback loops.



# Sample Actions for Improving the Usability of Data Quality Information throughout the Data Lifecycle

|  | Data Center  | Science Team  |
|--|--|---|
| <b>Capturing</b><br>Data Quality Information       | <b>request documentation from investigators</b> on the extent of error introduced into data products ...   | develop capabilities for investigators to <b>describe the extent of error</b> introduced ...  |
| <b>Describing</b><br>Data Quality Information      | <b>provide enough publicly available information</b> so users do not need to contact the data center   | <b>describe quality flags</b> in the data documentation and in the FAQs   |
| <b>Discovery of</b><br>Data Quality Information    | develop capabilities for users to <b>refine search query</b> results by selecting among choices of <b>quantifiable data quality criteria</b> , such as confidence levels ... | <b>identify quantifiable data quality criteria</b> , such as confidence levels and the values of quality flags, that can be used as criteria for refining search queries. |
| <b>Enabling Use of</b><br>Data Quality Information | provide users with a tool to <b>identify inputs, ..., that contributed to each pixel.</b>  | create tools to capture into a variable, <b>sensor inputs, ... that contributed to each pixel.</b>  |

# Sample Inventory of Current Implementations for Data Systems Integration

|   | Data Center -<br>Investigator<br>Communication | Metadata<br>Creation &<br>validation | Guidance<br>&<br>Instruction | Reference<br>/ Help<br>Desk |
|---|--|--------------------------------------|------------------------------|-----------------------------|
| Metadata Compliance<br>Checker                  |  | X                                    |                              |                             |
| Data Quality Guide<br>Document                  |  |                                      | X                            | X                           |
| Science Data Working<br>Group                   | X  |                                      |                              |                             |
| Data Quality Section in<br>Data Management Plan | X  |                                      | X                            |                             |
| Data Management Plan<br>/ Guidelines            | X  |                                      | X                            |                             |
| FAQ Development and<br>Analysis                 |  |                                      |                              | X                           |

# Implications for Science Practice



- Data quality raises the stakes for data contributions
  - Effort needed to measure, document, and disseminate data quality
  - Data quality measures identify positive and negative aspects of data
- Recognition for data and data quality contributions
  - Data quality review is a scientific contribution and includes design of measures, administration of review process, conducting reviews, documenting results, and enabling use of data quality information

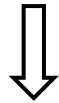
# Logical Progression for Planning Data Quality Assessments for Specific Uses

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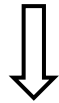
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- Plans to disseminate data should be justified by potential uses of the data



- Justification to disseminate data should include a description of the potential uses of the data



- Upon identifying a proposed use for data, plan to conduct a data quality assessment for the proposed uses of the data

# Recommendations for Communicating Data Quality Information

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- Each publicly accessible data set should describe its potential use
- Each claim for a potential use of data should be justified by a quality assessment
- Each data quality assessment should include a data quality indicator in data documentation or metadata
- Each data quality assessment indicator value should be defined in the data documentation or metadata
- Data users should cite data and describe their assessment of the data for the study conducted
- Update metadata and documentation to reflect reported data assessments



# Taking Action to Initiate, Curate, and Disseminate Data Quality Reviews

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- Archives, Repositories, and Data Centers
  - Invite user community to review data
  - Establish roles for user community members who review data
  - Include data review results when archiving data
  - Disseminate data review results with data
- Future Research Opportunities
  - Identifying “low-hanging fruit” solutions that could be reasonably executed in a relatively short time frame (current DQWG effort)
  - Identifying ways to standardize data quality practices and workflows
  - Identifying additional data quality challenges
  - Proposing new concepts to address challenges where existing solutions and best practices fall short